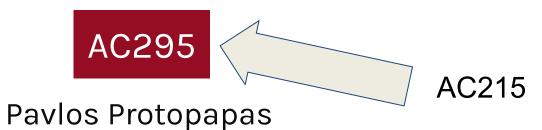
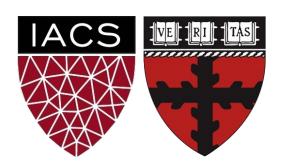
StyleCLIP - Project Outline

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Outline

- Project Scope
- Project Workflow
- Process Flow
- Data
- Models

Problem Definition

Pictures of various people exist but are normally pretty boring and we have little control over them. If you want to edit pictures in exciting ways, you have to learn Photoshop, which is quite difficult. There is no easy way to apply transformations to people in images and we'd like to be able to do so without much time or computational power.

Proposed Solution

We'll have an app on our phone that allows us to take or upload pictures. We will then be able to type in a description of the current picture and a description of the transformation we'd like to make. The app will then make that transformation to the image and return it. For example, we could take a picture of Pavlos and caption it "a face." If we added the transformation description "a bald face," nothing would happen.

Project Scope

Proof Of Concept (POC)

- Be able to run pretrained StyleCLIP model on our images
- Find additional images to train on (Celeb_a_HQ dataset)
- Finetune e4e model (used for StyleCLIP global) on new images
- Connect new model to CLIP and be able to generate new transformations

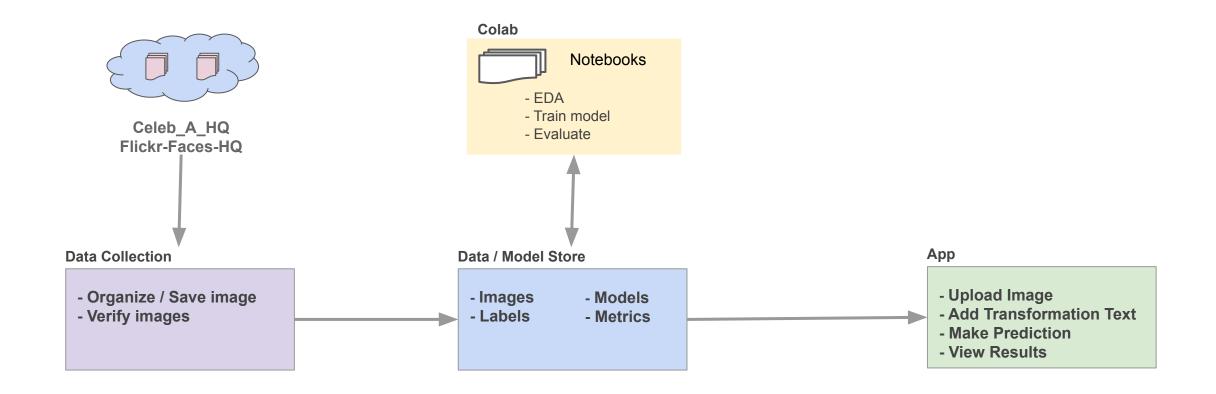
Prototype

- Create a mockup of screens to see how the app could look like
- Deploy StyleCLIP with finetuned e4e to Fast API to service model predictions as an API

Minimum Viable Product (MVP)

- Create containerized app to apply transformations to images
- API Server for uploading images and predicting using best model

Process Flow



Data - celeb_a_hq



















Data Details

- Total number of images: 30000
- Dimensions: (1024,1024)
- Attributes: 40 binary, 5 landmarks

Models

Encoding of images:

• Transfer learning, <u>e4e model</u>

Models - Training Results

```
Metrics for train, step 150
       d real loss = 0.6163466572761536
       d fake loss = 0.6397031545639038
       discriminator loss = 1.2560498714447021
       encoder_discriminator_loss = 0.7536979913711548
       total delta loss = 0.0
       loss id = 0.788224458694458
       id improve = -0.7882244782522321
       loss 12 = 0.14702948927879333
       loss lpips = 0.405124306678772
       loss = 0.9406110048294067
Metrics for train, step 200
       d real loss = 0.5907375812530518
       d fake loss = 0.5820201635360718
       discriminator loss = 1.1727577447891235
       encoder discriminator loss = 0.8309675455093384
       total delta loss = 0.0
       loss id = 0.6942740678787231
       id improve = -0.6942740976810455
       loss 12 = 0.17666828632354736
       loss lpips = 0.4156036972999573
       loss = 0.9393850564956665
```

